

04_modeling_opec_price_forecasting

December 2, 2025

1 Modeling OPEC Sentiment vs PP Prices

Predict next-month PP_EU using GPT comparison scores, hybrid index, FinBERT sentiment, and keyword densities.

```
[1]: from pathlib import Path
BASE_DIR = Path.cwd()
if BASE_DIR.name == 'notebooks':
    BASE_DIR = BASE_DIR.parent

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import mean_absolute_error
from sklearn.ensemble import RandomForestRegressor
from xgboost import XGBRegressor

sns.set_style('whitegrid')

data_path = BASE_DIR / 'data' / 'processed' / 'master_opec_price_model_dataset.
˓→csv'
df = pd.read_csv(data_path)
df['date'] = pd.to_datetime(df['date'])
df = df.sort_values('date').reset_index(drop=True)
df.head()
```

```
[1]:      date  comparison_score  hybrid_index  finbert_sentiment  supply_up \
0  2019-01-31           0.0          0.0        0.110262  0.004327
1  2019-02-28          -0.7         -0.7        0.046981  0.002033
2  2019-03-31          -0.7         -1.4        0.061729  0.008493
3  2019-04-30          -0.6         -2.0       -0.239369  0.000517
4  2019-05-31          -0.3         -2.3       -0.102873  0.003960

      supply_down  demand_up  demand_down  price_up  price_down  PP_EU  Brent \
0     0.004103   0.002233   0.004360  0.000000   0.001540  1385.0  59.77
1     0.004222   0.000220   0.004228  0.000097   0.001873  1385.0  63.63
2     0.003438   0.003655   0.004500  0.000073   0.001865  1410.0  66.66
```

```

3    0.001615  0.000050    0.001818  0.000805    0.001073  1430.0  71.03
4    0.004860  0.002240    0.004933  0.000025    0.000148  1430.0  70.93

```

	WTI	NatGas	PP_EU_next_month	Brent_next_month
0	51.07	3.15	1385.0	63.63
1	54.53	2.69	1410.0	66.66
2	57.62	2.83	1430.0	71.03
3	63.22	2.62	1430.0	70.93
4	61.76	2.60	1430.0	63.35

1.1 Dataset overview

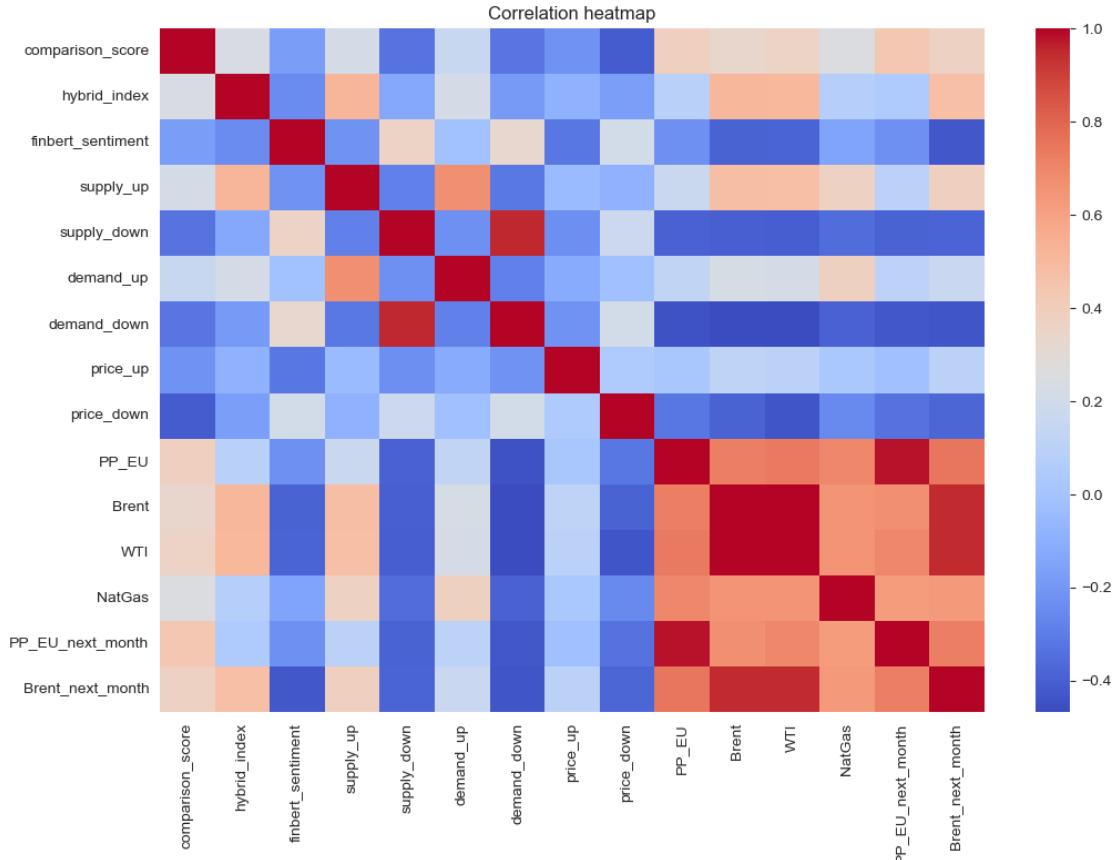
Columns include raw GPT comparison scores, cumulative hybrid index, FinBERT sentiment, keyword densities, and market prices aligned to month-end.

```
[2]: df.describe(include='all').T
```

	count	mean	min	\
date	80	2022-05-26 19:12:00	2019-01-31 00:00:00	
comparison_score	80.0	0.11625	-0.9	
hybrid_index	80.0	-0.59625	-10.8	
finbert_sentiment	80.0	-0.046661	-0.245568	
supply_up	80.0	0.00598	0.000517	
supply_down	80.0	0.002217	0.0002	
demand_up	80.0	0.002424	0.00005	
demand_down	80.0	0.002306	0.000265	
price_up	80.0	0.000221	0.0	
price_down	80.0	0.000401	0.0	
PP_EU	80.0	1565.125	995.0	
Brent	80.0	72.775875	27.29	
WTI	80.0	68.38275	17.42	
NatGas	80.0	3.35625	1.71	
PP_EU_next_month	80.0	1565.875	995.0	
Brent_next_month	80.0	72.829875	27.29	
	25%	50%	\	
date	2020-09-22 12:00:00	2022-05-15 12:00:00		
comparison_score	-0.45	0.35		
hybrid_index	-5.525	-0.75		
finbert_sentiment	-0.119719	-0.057719		
supply_up	0.003641	0.006189		
supply_down	0.001404	0.001944		
demand_up	0.001324	0.002092		
demand_down	0.001469	0.001999		
price_up	0.000025	0.000046		
price_down	0.000048	0.000112		
PP_EU	1385.0	1451.25		

Brent	63.855	73.555	
WTI	57.4775	70.375	
NatGas	2.3675	2.745	
PP_EU_next_month	1385.0	1451.25	
Brent_next_month	64.05	73.555	
	75%	max	std
date	2024-02-07 06:00:00	2025-09-30 00:00:00	NaN
comparison_score	0.6	0.8	0.563094
hybrid_index	5.05	9.3	5.840007
finbert_sentiment	0.027162	0.300448	0.114086
supply_up	0.008089	0.012422	0.002935
supply_down	0.002726	0.008483	0.001372
demand_up	0.003445	0.00717	0.001556
demand_down	0.002699	0.007567	0.00139
price_up	0.000456	0.000958	0.000294
price_down	0.00064	0.001873	0.000494
PP_EU	1722.5	2475.0	376.825703
Brent	83.27	118.14	18.021506
WTI	78.865	115.19	18.177425
NatGas	3.735	8.71	1.62051
PP_EU_next_month	1722.5	2475.0	376.522248
Brent_next_month	83.27	118.14	17.988483

```
[3]: plt.figure(figsize=(12, 8))
corr = df.corr(numeric_only=True)
sns.heatmap(corr, cmap='coolwarm', annot=False)
plt.title('Correlation heatmap')
plt.show()
```



1.2 Train/validation split

Use the earliest 80% of months for training and the latest 20% for validation to respect time order.

```
[4]: feature_cols = [
    'comparison_score', 'hybrid_index', 'finbert_sentiment',
    'supply_up', 'supply_down', 'demand_up', 'demand_down',
    'price_up', 'price_down', 'Brent', 'WTI', 'NatGas', 'PP_EU',
]

target = 'PP_EU_next_month'
split_idx = int(len(df) * 0.8)
X_train, X_val = df.loc[:split_idx - 1, feature_cols], df.loc[split_idx:, :]
y_train, y_val = df.loc[:split_idx - 1, target], df.loc[split_idx:, target]
dates_val = df.loc[y_val.index, 'date']

naive_pred = df.loc[y_val.index, 'PP_EU']
naive_mae = mean_absolute_error(y_val, naive_pred)
print('Naive MAE (predict current PP as next month):', round(naive_mae, 2))
```

Naive MAE (predict current PP as next month): 11.25

```
[5]: rf = RandomForestRegressor(n_estimators=300, random_state=42, min_samples_leaf=2)
rf.fit(X_train, y_train)
rf_pred = rf.predict(X_val)
rf_mae = mean_absolute_error(y_val, rf_pred)

xgb = XGBRegressor(
    random_state=42, n_estimators=400, learning_rate=0.05, max_depth=4,
    subsample=0.9, colsample_bytree=0.9
)
xgb.fit(X_train, y_train)
xgb_pred = xgb.predict(X_val)
xgb_mae = mean_absolute_error(y_val, xgb_pred)

results = pd.DataFrame(
    [
        ['Naive (current PP)', naive_mae],
        ['RandomForest', rf_mae],
        ['XGBoost', xgb_mae],
    ],
    columns=['model', 'mae']
)
results
```

```
[5]:      model      mae
0  Naive (current PP)  11.250000
1      RandomForest  23.632629
2        XGBoost   37.491753
```

```
[6]: val_df = pd.DataFrame({
    'date': dates_val,
    'actual': y_val.values,
    'rf_pred': rf_pred,
    'xgb_pred': xgb_pred,
})

plt.figure(figsize=(12, 6))
plt.plot(val_df['date'], val_df['actual'], label='Actual', linewidth=2)
plt.plot(val_df['date'], val_df['rf_pred'], label='RF pred', alpha=0.8)
plt.plot(val_df['date'], val_df['xgb_pred'], label='XGB pred', alpha=0.8)
plt.legend()
plt.title('Validation: next-month PP_EU')
plt.xlabel('Date')
plt.ylabel('EUR/t')
plt.xticks(rotation=45)
```

```
plt.tight_layout()  
plt.show()
```

